

# Package ‘LVMMCOR’

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**Title** A Latent Variable Model for Mixed Continuous and Ordinal Responses

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**Description** A model for mixed ordinal and continuous responses is presented where the heteroscedasticity of the variance of the continuous response is also modeled. In this model ordinal response can be dependent on the continuous response. The aim is to use an approach similar to that of Heckman (1978) for the joint modelling of the ordinal and continuous responses. With this model, the dependence between responses can be taken into account by the correlation between errors in the models for continuous and ordinal responses

**Depends** R (>= 2.14.0),nlme,MASS

**License** GPL (>= 2)

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LVMMCOR-package	<i>A Latent Variable Model for Mixed Continuous and Ordinal Responses.</i>
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## Description

A model for mixed ordinal and continuous responses is presented where the heteroscedasticity of the variance of the continuous response is also modeled.

## Details

Package: LVMMCOR  
 Type: Package  
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## Author(s)

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## References

Bahrami Samani, E., Ganjali, M. and Khodaddadi, A. (2008). A Latent Variable Model for Mixed Continuous and Ordinal Responses. *Journal of Statistical Theory and Applications*. 7(3):337-349.

## Examples

```
data("Bahrami")
gender<-Bahrami$ GENDER
age<-Bahrami$AGE
duration <-Bahrami$ DURATION
y<-Bahrami$ STEATOS
z<-Bahrami$ BMI
sbp<-Bahrami$ SBP
X=cbind(gender,age,duration ,sbp)
P<-lm(z~X)[[1]]
names(P)<-paste("Con_",names(P),sep="")
Q<-polr(factor(y)~X)[[1]]
names(Q)<-paste("Ord_",names(Q),sep="")
W=c(cor(y,z),polr(factor(y)~X)[[2]],var(z))
names(W)=c("Corr","cut_point1","cut_point2","Variance of Continous Response")
ini=c(P,Q,W)
```

```
p=5;  
q=4;  
LVMMCOR(ini, X=X, y=y, z=z, p=p, q=q)
```

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Bahrami

*Body Mass Index and Steatosis Data*

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### **Description**

The medical data set is obtained from an observational study in the Taleghani hospital in Tehran. These data record the Steatosis and BMI for 15 diabetic patients. Steatosis is the process describing the abnormal retention of lipids within a cell. It reflects an impair of the normal process of synthesis and breakdown of triglyceride fat. Excess lipid accumulates in vesicles that displace the cytoplasm. BMI is a statistical measure of the weight of body mass index. A person scaled height body mass index may be accurately calculated using any of the formulas such as  $BMI=w/H^2$  where  $W$  is weight and  $H$  is height. As explanatory variable for both Steatosis and BMI, the systolic blood pressure (SBP) is defined as the peak pressure in the arteries, which occurs near the beginning of the cardiac cycle. The normal rate, in adult humans, for systolic is near but less than 120 mmHg. As another explanatory variable, duration of diabet is an amount of time or a particular time interval which a person take diabet (a metabolic disorder characterized by high blood sugar and other signs). Two more explanatory variables are age and gender.

### **Usage**

```
data(Bahrami)
```

### **Format**

A data frame with 15 observations on the following 6 variables: AGE, DURATION, SBP, STEATOS, BMI, GENDER

### **Source**

The medical data set is obtained from an observational study in the Taleghani hospital in Tehran.

### **References**

Bahrami Samani, E., Ganjali, M. and Khodaddadi, A. (2008). A Latent Variable Model for Mixed Continuous and Ordinal Responses. *Journal of Statistical Theory and Applications*. 7(3):337-349.

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LVMMCOR

*A Latent Variable Model for Mixed Continuous and Ordinal Responses.*


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### Description

A model for mixed ordinal and continuous responses is presented where the heteroscedasticity of the variance of the continuous response is also modeled. In this model ordinal response can be dependent on the continuous response. The aim is to use an approach similar to that of Heckman (1978) for the joint modelling of the ordinal and continuous responses. With this model, the dependence between responses can be taken into account by the correlation between errors in the models for continuous and ordinal responses.

### Usage

```
LVMMCOR(ini = NA, X, y, z, p, q, ...)
```

### Arguments

ini	Initial values
X	Design matrix
z	Continuous responses
y	Ordinal responses with three levels
p	Order of dimension of continuous responses
q	Order of dimension of ordinal responses
...	Other arguments

### Details

Models for LVMMCOR are specified symbolically. A typical model has the form `response1 ~ terms` and `response2 ~ terms` where `response1` and `response2` are the (numeric) ordinal and continuous responses vector and `terms` is a series of terms which specifies a linear predictor for responses. A terms specification of the form `first + second` indicates all the terms in `first` together with all the terms in `second` with duplicates removed. A specification of the form `first:second` indicates the set of terms obtained by taking the interactions of all terms in `first` with all terms in `second`. The specification `first*second` indicates the cross of `first` and `second`. This is the same as `first + second + first:second`.

### Value

Continuous Response	Coefficient of continuous response
Variance of Continuous Response	Variance of continuous response
Ordinal Response	Coefficient of ordinal response

Cut points	Cut points for ordinal response
Correlation	Coefficient of continuous response
Hessian	Hessian matrix
convergence	An integer code. 0 indicates successful convergence

**Note**

Supported by Shahid Beheshti University

**Author(s)**

Bahrami Samani and Nourallah Tazikeh Miyandarreh

**References**

Bahrami Samani, E., Ganjali, M. and Khodaddadi, A. (2008). A Latent Variable Model for Mixed Continuous and Ordinal Responses. *Journal of Statistical Theory and Applications*. 7(3):337-349.

**See Also**

[nlminb,fdHess](#)

**Examples**

```

data("Bahrami")
gender<-Bahrami$ GENDER
age<-Bahrami$AGE
duration <-Bahrami$ DURATION
y<-Bahrami$ STEATOS
z<-Bahrami$ BMI
sbp<-Bahrami$ SBP
X=cbind(gender,age,duration ,sbp)
P<-lm(z~X)[[1]]
names(P)<-paste("Con_",names(P),sep="")
Q<-polr(factor(y)~X)[[1]]
names(Q)<-paste("Ord_",names(Q),sep="")
W=c(cor(y,z),polr(factor(y)~X)[[2]],var(z))
names(W)=c("Corr","cut_point1","cut_point2","Variance of Continous Response")
ini=c(P,Q,W)
p=5;
q=4;
LVMMCOR(ini,X=X,y=y,z=z,p=p,q=q)

## The function is currently defined as
structure(function (x, ...)
UseMethod("LVMMCOR"), class = "LVMMCOR")

```

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LVMMCOR.default	<i>A Latent Variable Model for Mixed Continuous and Ordinal Responses.</i>
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### Description

A model for mixed ordinal and continuous responses is presented where the heteroscedasticity of the variance of the continuous response is also modeled. In this model ordinal response can be dependent on the continuous response. The aim is to use an approach similar to that of Heckman (1978) for the joint modelling of the ordinal and continuous responses. With this model, the dependence between responses can be taken into account by the correlation between errors in the models for continuous and ordinal responses.

### Usage

```
## Default S3 method:
LVMMCOR(ini = NA, X, y, z, p, q, ...)
```

### Arguments

ini	Initial values
X	Design matrix
z	Continuous responses
y	Ordinal responses with three Levels
p	Order of dimension of continuous responses
q	Order of dimension of ordinal responses
...	Other arguments

### Details

Models for LVMMCOR are specified symbolically. A typical model has the form `response1 ~ terms` and `response2 ~ terms` where `response1` and `response2` are the (numeric) ordinal and continuous responses vector and `terms` is a series of terms which specifies a linear predictor for responses. A terms specification of the form `first + second` indicates all the terms in `first` together with all the terms in `second` with duplicates removed. A specification of the form `first:second` indicates the set of terms obtained by taking the interactions of all terms in `first` with all terms in `second`. The specification `first*second` indicates the cross of `first` and `second`. This is the same as `first + second + first:second`.

### Value

Continuous Response	
	Coefficient of continuous response
Variance of Continuous Response	
	Variance of continuous response

Ordinal response	Coefficient of ordinal response
Cut points	Cut points for ordinal response
Correlation	Coefficient of continuous response
Hessian	Hessian matrix
convergence	An integer code. 0 indicates successful convergence.

**Note**

Supportted by Shahid Beheshti University

**Author(s)**

Bahrami Samani and Nourallah Tazikeh Miyandarreh

**References**

Bahrami Samani, E., Ganjali, M. and Khodaddadi, A. (2008). A Latent Variable Model for Mixed Continuous and Ordinal Responses. *Journal of Statistical Theory and Applications*. 7(3):337-349.

**See Also**

[nlminb,fdHess](#)

**Examples**

```
function (ini = NA, X, y, z, p, q, ...)
{
  options(warn = -1)
  f <- function(ini, X, y, z, p, q) {
    X = cbind(1, X)
    y <- as.vector(y)
    z <- as.vector(z)
    ini <- as.vector(ini)
    X <- as.matrix(X)
    n = nrow(X)
    muz = muy = muygivenzx = q2 = q1 = l1 = l2 = l3 = muygivenzx = as.vector(0)
    sez <- exp(ini[p + q + 4])
    seygivenzx <- (1 - (ini[p + q + 1])^2)
    for (i in 1:n) {
      muz[i] <- as.numeric(t(ini[1:p]) %*% X[i, ])
      muy[i] <- as.numeric(t(ini[(p + 1):(p + q)]) %*%
        X[i, -1])
      muygivenzx[i] <- muy[i] + (ini[p + q + 1] * (z[i] -
        muz[i]))/sez
      q1[i] <- (ini[p + q + 2] - muygivenzx[i])/sqrt(seygivenzx)
      q2[i] <- (ini[p + q + 3] - muygivenzx[i])/sqrt(seygivenzx)
      l1[i] <- log(pnorm(q1[i])) + log(dnorm(z[i], muz[i],
        sez))
      l2[i] <- log(pnorm(q2[i]) - pnorm(q1[i])) + log(dnorm(z[i],
        muz[i], sez))
    }
  }
}
```

```

      l3[i] <- log(1 - pnorm(q2[i])) + log(dnorm(z[i],
        muz[i], sez))
    }
    data0 <- cbind(y, l1)
    data1 <- cbind(y, l2)
    data2 <- cbind(y, l3)
    data0[data0[, 1] == 1, 2] <- 0
    data0[data0[, 1] == 2, 2] <- 0
    data1[data1[, 1] == 0, 2] <- 0
    data1[data1[, 1] == 2, 2] <- 0
    data2[data2[, 1] == 0, 2] <- 0
    data2[data2[, 1] == 1, 2] <- 0
    t0 <- sum(data0[, 2])
    t1 <- sum(data1[, 2])
    t2 <- sum(data2[, 2])
    t <- (c(t0, t1, t2))
    Tfinal <- sum(t)
    return(-Tfinal)
  }
n = nlminb(ini, f, X = X, y = y, z = z, p = p, q = q, lower = c(rep(-Inf,
  length(ini)), -0.999, -Inf, -Inf, 0), upper = c(rep(Inf,
  length(ini)), 0.999, Inf, Inf, Inf), hessian = T)
h = fdHess(n$par, f, z = z, y = y, X, p, q)
h1 = h$Hessian
ih = ginv(h1)
se = sqrt(abs(diag(ih)))
n$Hessian <- h1
n$p <- p
n$q <- q
n$se <- as.vector(se)
n$call <- match.call()
class(n) <- "LVMMCOR"
object = n
Co.Re <- data.frame(Parameter = object$par[1:p], S.E = object$se[1:p],
  `Confidence Interval` = paste("(", round(object$par[1:p] -
    2 * object$se[1:p], 3), ",", round(object$par[1:p] +
    2 * object$se[1:p], 3), ")", sep = ""))
Or.Re <- data.frame(Parameter = object$par[(p + 1):(p + q)],
  S.E = object$se[(p + 1):(p + q)], `Confidence Interval` = paste("(",
    round(object$par[(p + 1):(p + q)] - 2 * object$se[(p +
    1):(p + q)], 3), ",", round(object$par[(p + 1):(p +
    q)] + 2 * object$se[(p + 1):(p + q)], 3), ")",
    sep = ""))
Cut.P <- data.frame(Parameter = object$par[(p + q + 2):(p +
  q + 3)], S.E = object$se[(p + q + 2):(p + q + 3)], `Confidence Interval` = paste("(",
    round(object$par[(p + q + 2):(p + q + 3)] - 2 * object$se[(p +
    q + 2):(p + q + 3)], 3), ",", round(object$par[(p +
    q + 2):(p + q + 3)] + 2 * object$se[(p + q + 2):(p +
    q + 3)], 3), ")", sep = ""))
Cor <- data.frame(Parameter = object$par[p + q + 1], S.E = object$se[p +
  q + 1], `Confidence Interval` = paste("(", round(object$par[p +
  q + 1] - 2 * object$se[p + q + 1], 3), ",", round(object$par[p +
  q + 1] + 2 * object$se[p + q + 1], 3), ")", sep = ""))

```



```
Var <- data.frame(Parameter = object$par[p + q + 4], S.E = object$se[p +
  q + 4], `Confidence Interval` = paste("(", round(object$par[p +
  q + 4] - 2 * object$se[p + q + 4], 3), ",", round(object$par[p +
  q + 4] + 2 * object$se[p + q + 4], 3), ")", sep = ""))
row.names(Cut.P) <- c("cut point1", "cut point2")
res <- list(call = object$call, `Continuous Response` = Co.Re,
  `Variance Of Countinous Response` = Var, `Ordinal Response` = Or.Re,
  `Cut points` = Cut.P, Correlation = Cor)
res$Hessian <- h1
res$convergence <- n$convergence
res$call <- match.call()
class(res) <- "LVMMCOR"
res
}
```

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